

AN ARTIFICIAL NEURAL NETWORK  
APPROACH FOR CLASSIFICATION AND  
PREDICTION OF LOW PRESSURE  
CHEMICAL VAPOR DEPOSITION  $\text{Si}_x\text{N}_y$  ISFET  
PH SENSOR DRIFT

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## **SUPERVISOR'S DECLARATION**

I hereby declare that I have checked this thesis and in my opinion, this thesis is adequate in terms of scope and quality for the award of the degree of Master of Science

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## **STUDENT'S DECLARATION**

I hereby declare that the work in this thesis is based on my original work except for quotations and citations which have been duly acknowledged. I also declare that it has not been previously or concurrently submitted for any other degree at Universiti Malaysia Pahang or any other institutions.

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## ABSTRAK

Sensor bagi transistor bersifat ion memilih kesan-lapangan (ISFET) adalah sensor lapangan dimana ion di dalam sesuatu sampel media mengalami pelbagai persekitaran yang mempengaruhi tindakbalas molekul sedida dan bertumpu di lapisan gerbang oksida. Satu perubahan dari segi caj elektrik berlaku dan memberi kesan terhadap konduktans dalam saluran ISFET. Akibatnya, perubahan konduktans dalam sumber dan saluran menghasilkan isyarat elektrik. Masalah yang biasa berlaku ialah berlakunya hanyutan voltan semasa isyarat elektrik beransur berubah dengan bebas daripada sampel yang diukur. Objektif utama kajian adalah untuk membina satu model berdasarkan jaringan neutral buatan yang boleh digunapakai untuk mengklasifikasi dan meramal kesilapan serta melaksanakan pampasan drift dalam sensor pemendapan wap kimia  $\text{Si}_x\text{N}_y$  ISFET pH. Model yang sedemikian boleh digunakan untuk menghadapi masalah hanyutan voltan yang biasanya wujud dalam sensor kimia. Tiga unit sensor ISFET telah digunakan untuk menentukan tiga jenis larutan penampapan pH iaitu pH 4, pH 7 dan pH 10. Rangkaian neural buatan telah digunakan untuk membina satu model kotak hitam berbilang input dan output data ISFET. Satu nilai peratusan ketepatan telah digunakan untuk menaksir prestasi model dari segi pengkelasan, sementara min ralat kuadrat (MSE) dan parameter koefisien penentuan ( $R^2$ ) digunakan bagi menentukan model terbaik dalam meramalkan kesilapan sensor ISFET. Dari segi struktur model dalam aspek klasifikasi, Pengenalan Corak Rangkaian Neural (PATTERNET) terbukti berfungsi lebih efisien daripada Rangkaian Fungsi Fitting (FITNET) dengan ketepatan 100%. Konfigurasi rangkaian PATTERNET dengan dwi-lapisan 30 nod pada lapisan tersembunyi pertama dan 3 nod pada lapisan tersembunyi kedua, mencapai keputusan terbaik. Dari segi ramalan, model NARX-BR, dengan 75 kelewatan, merupakan model yang cekap dalam meramalkan kesilapan set data ISFET. Nilai  $\text{MSE} = 4.881\text{e}^{-5}$  dan  $R^2 = 0.99930$  bagi model NARX-BR menunjukkan model mampu meramalkan ralat. Pampasan drift telah dilaksanakan dan isu kehanyutan dalam sensor ISFET telah berjaya diatasi. Kajian ini menunjukkan potensi yang ketara dalam pembangunan rangkaian neural buatan untuk menghalang isu hanyutan voltan dalam sensor tekanan rendah wap kimia  $\text{Si}_x\text{N}_y$  ISFET pH.

## ABSTRACT

A sensor in an ion selective field-effect transistor (ISFET) is a field sensor in which ions within a sample media undergo multiple environments affecting reactions occurring molecules to concentrate at the gate oxide layer. A change in electrical charge occurs and it affects the conductance in the ISFET channels. Consequently, the change in conductance within the source and the drain produce an electrical signal. The most common problem is the occurrence of drift when the electrical signal output gradually changes independent of the measured sample. The primary objective of the present study is to investigate a reliable model based on artificial neural network to classify and forecast errors as well as to implement the drift compensation in low-pressure chemical vapour deposition  $\text{Si}_x\text{N}_y$  ISFET pH sensor. Such model could be used to encounter voltage drift problems that usually exist in chemical sensors. Three units of ISFET sensors were used to calibrate with three types of pH buffer solutions that are pH 4, pH 7 and pH 10. Artificial neural networks were applied to construct a black-box with multiple inputs and outputs models of the ISFET data. A percentage accuracy value was used to assess the model's performances in terms of classification, while the mean squared error (MSE) and the coefficient of determination ( $R^2$ ) parameter were used to determine the best models in predicting errors in the ISFET sensors. On the model's structure in classification, Pattern Recognition Neural Network (PATTERNNET) proved to perform better than Function Fitting (FITNET) networks with 100% accuracy. The network configuration in PATTERNNET, a dual-layered network with 30 nodes on the first hidden layer and 3 nodes on the second hidden layer achieved the best results. As for prediction, the NARX-BR model with 75 delays produced an efficient model in predicting errors in ISFET data set. The value of  $\text{MSE} = 4.8814\text{e}^{-5}$  and  $R^2 = 0.99930$  for the NARX-BR model showed that the model is capable of predicting errors. The drift compensation was applied and the drift issues in the ISFET sensors were successfully resolved. The present study demonstrates significant potentials in the development of artificial neural networks to stave off voltage drift issues in ISFET low-pressure chemical vapour deposition  $\text{Si}_x\text{N}_y$  ISFET pH sensor.

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## **LIST OF ABBREVIATIONS**

AANN	Auto-Associative Neural Network
ANN	Artificial Neural Network
APS	Active Pixel Sensor
BPN	Backpropagation Neural Network
BR	Bayesian Regularization
CBPN	Cascade Forward Back Propagation Neural Network
CMOS	Complementary Metal-Oxide-Semiconductor
CSV	Comma-Separated Values
CTAT	Complementary To Absolute Temperature
DNA	Deoxyribonucleic Acid
DTDNN	Distributed Time-Delay Neural Network
FITNET	Function Fitting Neural Network
IGFET	Insulated-Gate Field-Effect Transistor
ISE	Ion-Selective Electrode
ISFET	Ion-Sensitive Field-Effect Transistor
LM	Levenberg-Marquardt
LPCVD	Low Pressure Chemical Vapor Deposition
MLP	Multilayer Perceptron
MOSFET	Metal-Oxide-Semiconductor Field-Effect Transistor
MSE	Mean Squared Error
MSPCA	Multiscale Principle Component Analysis
NAR	Nonlinear Autoregressive
NARX	Nonlinear Autoregressive With External Input
NMOS	Negative Channel Metal-Oxide-Semiconductor
PATTERNNET	Pattern Recognition Neural Network
PWM	Pulse With Modulation
REFET	Reference Field-Effect Transistor
SOM	Self-Organizing Map
TDNN	Time Delay Neural Network

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